

this time: decision theory
 next time: unknown
 time: $\theta \in (0, 1)$

read: \textcircled{J} (ch 1, 2; Appendixes A, B) new:

AMS 206
 23 Jun 18

\textcircled{J} ch 3; \textcircled{G} (ch. 1)
 new: G ch. 2

$\textcircled{1}$
 ELISA
 case study
 (continued)

posterior
 odds
 ratio
 in favor
 of A given \oplus

$$= 0 = \left(\frac{1}{99}\right) \left(\frac{47.5}{1}\right) = \frac{95}{198}$$

$$\textcircled{O} \left(\frac{p}{1-p} \right) \leftrightarrow$$

$$p = \frac{0}{1+0} = \frac{95}{198} = \frac{1 + \frac{95}{198}}{293} = \frac{95}{293} \approx 0.32$$

III compute $P(A | \oplus)$
 directly with
 Bayes's Thm.

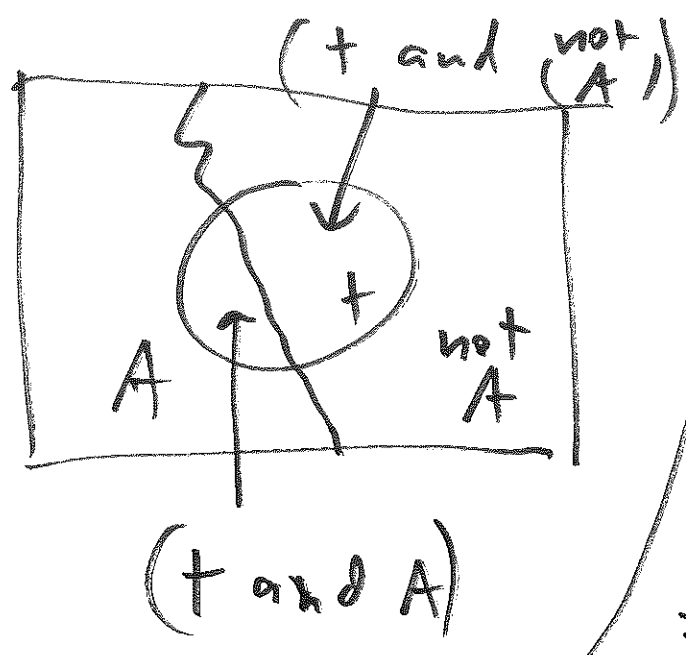
$$P(A|+) = \frac{P(\check{A}) \cdot P(+|A)}{P(+)} \quad (2)$$

$P(+)$ ← ?

DV Lindley
"extending
the conversation"

data

$$P(+)=P(+ \text{ and } A) +$$



$P(+ \text{ and } (\overset{\text{not}}{A})) \{A, \text{not } A\}$

$P(+ \text{ and } A)$
 $= P(A) \cdot P(+|A)$ dimension
by partitioning on another

Law of total probability

$$P(+ \text{ and } (\overset{\text{not}}{A})) = P(\overset{\text{not}}{A}) \cdot P(+|\overset{\text{not}}{A})$$

$$P(A|+) = \frac{P(A) \cdot P(+|A)}{P(A) \cdot P(+|A) + P(\overset{\text{not}}{A}) \cdot P(+|\overset{\text{not}}{A})}$$

$$P(A) \cdot P(+|A) + P(\overset{\text{not}}{A}) \cdot P(+|\overset{\text{not}}{A})$$

3

$$\frac{(.01)(.95)}{}$$

$$\frac{(.01)(.95)}{.95} + \frac{(.99)(1-.98)}{.02}$$

$P(A) = .01$
 $P(+|A) = .95$
 $P(-|A) = .98$

$$= \frac{95 + 198}{293} = 0.32$$

decision theory

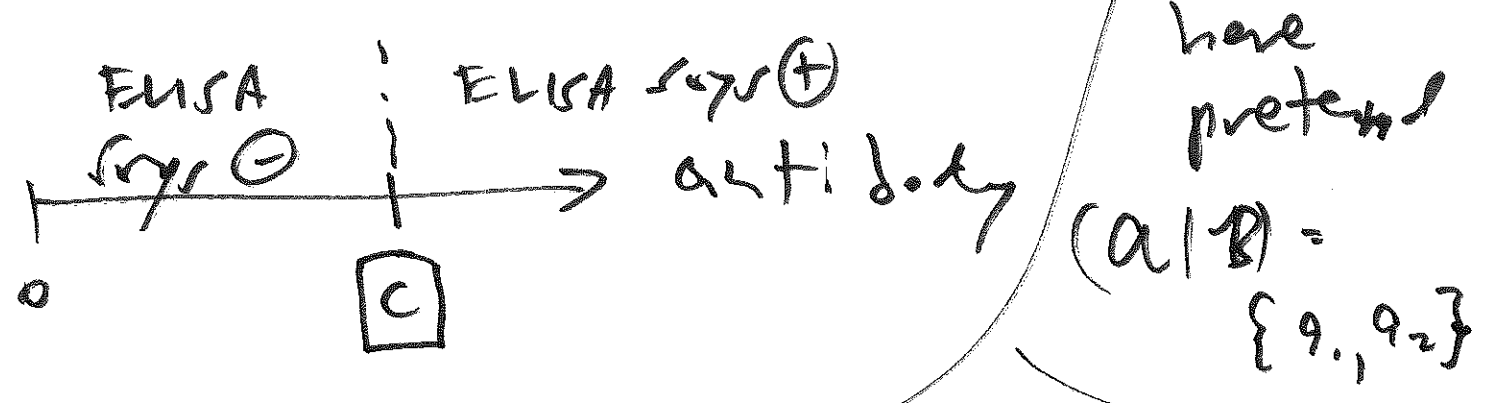
possible actions

action space (a|b)

$a_1 =$ (run ELISA: if ELISA says \oplus , declare (really is \oplus); if ELISA says \ominus , declare (really is \ominus))

$a_3 = (\text{run WB}; \text{if WB says } \oplus, \text{ declare } \textcircled{A}; \text{if WB says } \ominus, \text{ declare } (\text{not } \textcircled{A})) \textcircled{4}$

$a_2 = (\text{run ELISA: if } \textcircled{+} \text{, run WB: if WB } \oplus \rightarrow \textcircled{A}; \text{if } \ominus \rightarrow (\text{not } \textcircled{A}); \text{if ELISA } \ominus, \text{ declare } (\text{not } \textcircled{A}))$



we now have to specify a utility function $U(a, \theta | B) \in \mathbb{R}^1$

↑ action ↑ unknown ←

wlog assume big U better than small U (5)

$U(a, \theta | B) = f(\text{costs, benefits})$
of action a if θ is the

if θ were known, optimal a would maximize

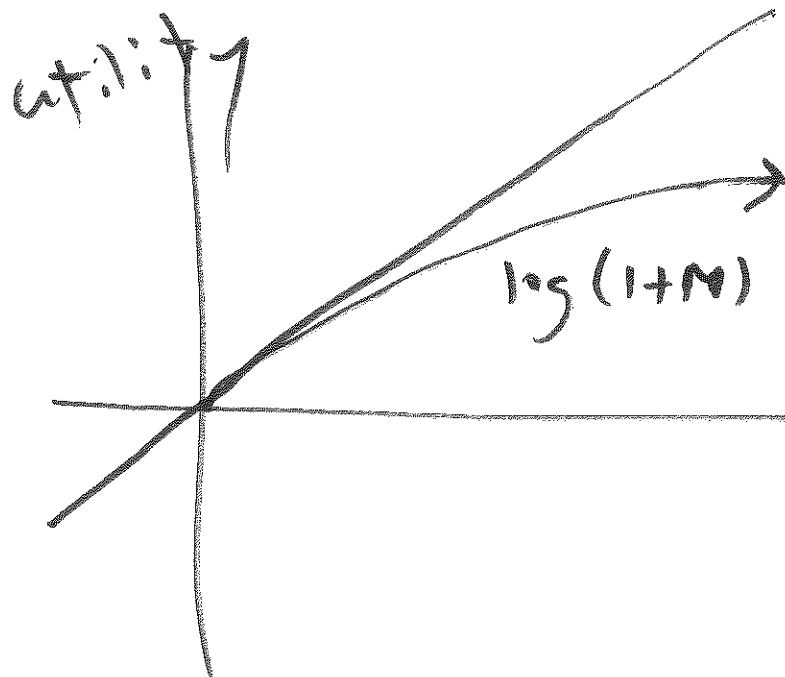
$U(a, \theta | B)$
 $a^* = \operatorname{argmax}_{a \in A(B)} U(a, \theta | B)$
optimal a

But can't do this because we don't know θ

utility:

Daniel Bernoulli
(1720)

6



utility = money

$$U(a, \theta | B) \sim \text{money}$$

money
M

$$U(M) = \log(1+M)$$

$U(a, \theta | B) = f(\text{costs, benefits})$
 from set a
 if unknown $= \theta$

$$= \boxed{g(\text{costs})} + h(\text{benefits})$$

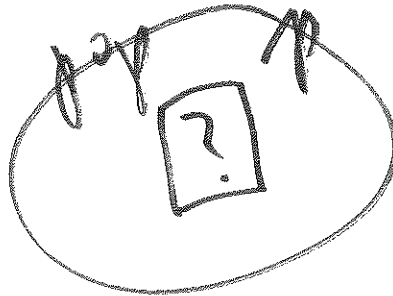
↑
monetary

prefer
 a_2 to a_1
 if

$$\underline{\underline{+0.0197 L_{II}}} - \underline{\underline{.0205 L_I}}$$

$$- .0293 C_2 > 0$$

↑

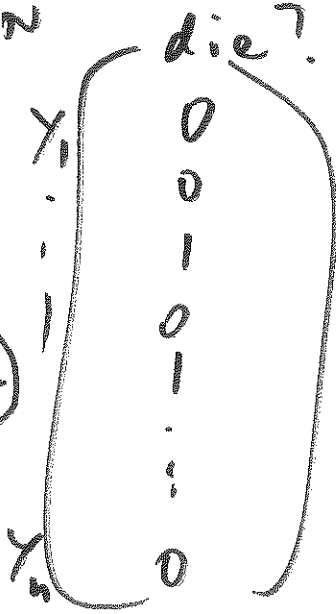
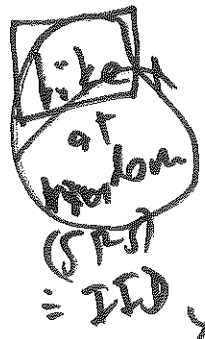
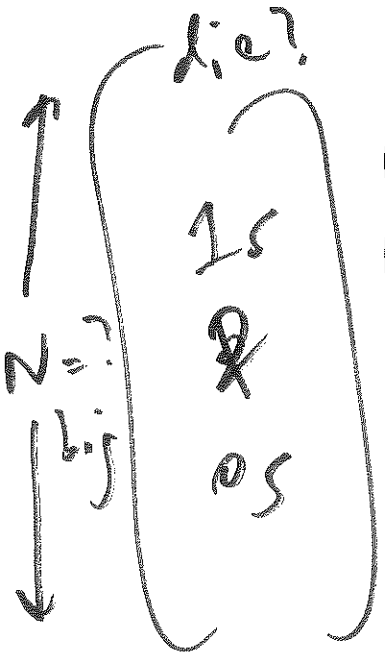


stat. inf.

Sample of the observed AMI patients

$$1 = Y$$

$$0 = N$$



$n = 400$

$$\sum_{i=1}^n Y_i = 72$$

$$P_F(|\bar{y} - \theta| \leq .01) = ?$$

high

mean $\theta = ?$

fixed unknown scalar

$$\bar{y} = \frac{1}{n} \sum_{i=1}^n Y_i = \frac{\sum_{i=1}^n Y_i}{n}$$

$$= \frac{72}{400} = 0.18 = 18\%$$

inference:

$\bar{y} = 18\%$ is a good guess (estimate) of θ ; but how good?

$$D(\text{data}) = \underline{y} = (Y_1, \dots, Y_n)$$